



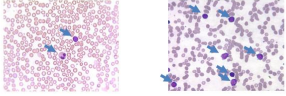
# White Blood Cell Differential Counting via Tiny YOLO

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## Motivation

- White blood cell (WBC) density and relative abundances are quantitative indications of our immune system state<sup>[1]</sup>



Healthy Patient<sup>[2]</sup> Leukemic Patient<sup>[2]</sup>

- Current methods to count WBCs are done manually or via expensive flow cytometers<sup>[3]</sup>
- Goal: Automate counting and differentiation of WBCs in microscopic blood smears**

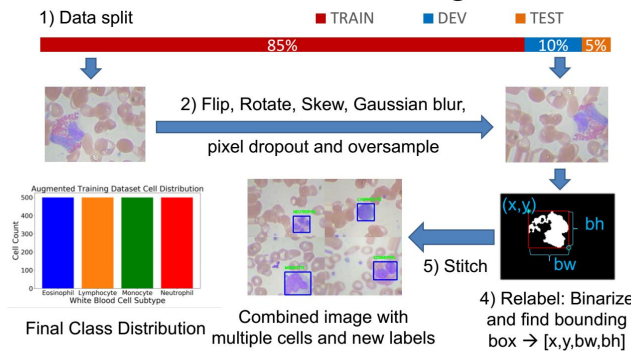
## Dataset

- There are 4 types of white blood cells that we classify and count (left)
- 348 Original 640 x 480 Images of single WBC surrounded by red blood cells<sup>[4]</sup>
- Skewed distribution of cells with many more neutrophils than others (bottom)
- Cell types pre-labeled by folder separation

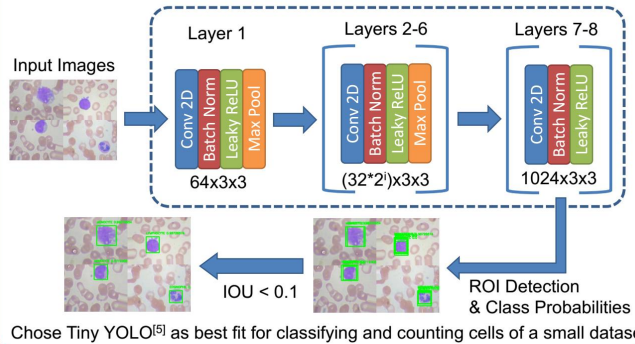
Original Dataset Cell Distribution

White Blood Cell Subtype

## Data Pre-Processing



## Tiny YOLO Model

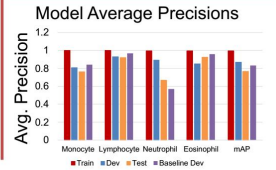
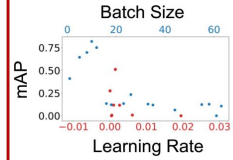
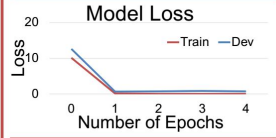


## Results

### Hyperparameter Sweep

- Course grid search with 2000 training images and 200 validation images and fine random search with all train images
- Swept: Learning rate, batch size, momentum, etc.

	mAP	Multicell?
Baseline	0.835	No
Ours	0.871	Yes



## Conclusion

- Made more useful WBC classification model to count cells
- Differentiated count of WBCs with 0.82 mAP on test set

## Future Works

- Train the network with a bigger dataset
- Train on images of blood smears containing multiple cells
- Add regularization to reduce overfitting to training set

## References

- [1] Blumenreich. Clinical Methods. 1990.
- [2] Carleton. Hematology of Leukemia. 2017.
- [3] Peng. Journal of Lab Hematology. 2007.
- [4] Mooney. Kaggle. 2018.
- [5] Redmon. CVPR. 2016.